- ¹ Title: Tree height and hydraulic traits shape growth responses across droughts in a temperate broadleaf
- 2 forest

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22 Summary

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- As climate change is driving increased drought frequency and severity in many forested regions around
 the world, mechanistic understanding of the factors conferring drought resistance in trees is
 increasingly important. The dendrochronological record provides a window through which we can
 understand how tree size and species' traits shape tree growth responses during droughts.
- We analyzed tree-ring records for twelve species that comprise 97% of the woody productivity of the 25.6-ha ForestGEO plot in a broadleaf deciduous forest of northern Virginia (USA) to test hypotheses on how tree height, microenvironment characteristics, and species' traits shaped drought responses across the three strongest regional droughts over a 60-year period (1950 2009).
 - Individual-level drought resistance decreased with tree height, which was strongly correlated with
 crown exposure. The potentially greater rooting volume of larger trees did not confer an advantage in
 sites with low topographic wetness index. Resistance was greater among species whose leaves
 experienced less shrinkage upon desiccation and lost turgor (wilted) at more negative water potentials.
- We conclude that tree height and hydraulic traits influence growth responses during drought, as recorded in the tree-ring record spanning historical droughts. Thus, these factors can be useful for predicting future drought responses under climate change.
- 38 Key words: annual growth; crown exposure; drought; Forest Global Earth Observatory (ForestGEO); leaf
- by hydraulic traits; temperate broadleaf deciduous forest; tree height; tree-ring

40 Introduction

- Forests play a critical global role in climate regulation (Bonan, 2008), yet there remains enormous uncertainty as to how the terrestrial carbon sink, which is dominated by forests, will respond to climate change (Friedlingstein et al., 2006). An important aspect of this uncertainty lies with physiological responses 43 of trees to drought (Kennedy et al., 2019). In many forested regions around the world, the risk of severe 44 drought is increasing (Trenberth et al., 2014; Dai et al., 2018), often despite increasing precipitation (Intergovernmental Panel on Climate Change, 2015; Cook et al., 2015). Droughts, intensified by climate change, have been affecting forests worldwide and are expected to continue as one of the most important drivers of forest change in the future (Allen et al., 2010, 2015). Understanding forest responses to drought requires elucidation of how tree size, microenvironment, and species' traits jointly influence individual-level drought resistance, and the extent to which their influence is consistent across droughts. However, it has proven difficult to resolve the many factors affecting tree growth during drought with available forest census data, which only rarely captures extreme drought, and with tree-ring records, which capture multiple droughts but rarely consider the roles of tree size and microenvironment. 53 Many studies have shown that within species, large trees tend to be more affected by drought. Greater growth reductions for larger trees was first shown on a global scale by Bennett et al. (2015), and subsequent studies have reinforced this finding (e.g., Stovall et al. (2019); Hacket-Pain et al. (2016)). It has yet to be resolved which of several potential underlying mechanisms most strongly shape size trends in drought 57 response. First, tree height may be a primary driver. Taller trees face the biophysical challenge of lifting water greater distances against the effects of gravity and friction (McDowell et al., 2011; McDowell and Allen, 2015; Ryan et al., 2006; Couvreur et al., 2018). Vertical gradients in stem and leaf traits-including smaller and thicker leaves (higher leaf mass per area, LMA), greater resistance to hydraulic dysfunction (i.e., more 61 negative water potential at 50% loss of hydraulic conductivity, more negative P50), and lower hydraulic 62 conductivity at greater heights (Couvreur et al., 2018; Koike et al., 2001; McDowell et al., 2011)-enable trees 63 to become tall (Couvreur et al., 2018). Indeed, tall trees require xylem of greater hydraulic efficiency, such that xylem conduit diameters are wider in the basal portions of taller trees, both within and across species (Olson et al., 2018; Liu et al., 2019), and throughout the conductive systems of angiosperms (Zak et al. 2010, Olson et al. 2014,2018). Wider xylem conduits plausibly make large trees more vulnerable to embolism during drought (Olson et al., 2018), and traits conducive to efficient water transport may also lead to poor ability to recover from or re-route water around embolisms (Roskilly et al., 2019). Larger trees may also have lower drought resistance because their crowns tend to occupy more exposed canopy positions, where they experience higher solar radiation, greater wind speeds, and lower relative humidity (REFS-KAT). 71 Subcanopy trees tend to fare better specifically due to the benefits of a buffered environment (Pretzsch et al., 2018). Counteracting the liabilities of associated with tall height, large trees tend to have larger root systems, which potentially counteracts some of the biophysical challenges they face by allowing greater access to water. 74 Finally, tree size-related responses to drought can be modified by species' traits and their distribution across size classes (Meakem et al., 2018; Liu et al., 2019). Understanding the mechanisms driving the greater relative growth reductions of larger trees during drought will require sorting out the interactive effects of 77 height and associated exposure, root water access, and species' traits. 78
- Debates have also arisen regarding the traits influencing tree growth responses to drought. Studies in temperate broadleaf forests have observed that ring-porous species showing higher drought tolerance than diffuse-porous species (Friedrichs et al., 2009; Elliott et al., 2015; Kannenberg et al., 2019), but this

distinction would not hold in the global context (Wheeler et al. 2007, Olson et al. 2020) and does not resolve differences among the many species within each category. Commonly-measured traits including wood density 83 and leaf mass per area (LMA) have been linked to drought responses in some temperate deciduous forests (Abrams, 1990; Guerfel et al., 2009; Hoffmann et al., 2011; Martin-Benito and Pederson, 2015) and other 85 forest biomes around the world (Greenwood et al., 2017). However, in other cases these traits could not explain drought tolerance (Maréchaux et al., 2019), or the direction of response was not always consistent. 87 For instance, higher wood density has been associated with greater drought resistance at a global scale (Greenwood et al., 2017), but it correlated negatively with tree performance during drought in a broadleaf deciduous forest in the southeastern United States (Hoffmann et al., 2011). Thus, the perceived influence of these traits on drought resistance may actually reflect indirect correlations with other traits that more 91 directly drive drought responses (Hoffmann et al., 2011). Recent work has shown a great potential for hydraulic traits to predict growth and mortality responses. Hydraulic traits including water potentials at which percent loss of conductivity surpass a certain threshold (P50, P80, P88) and hydraulic safety margin 94 correlate with drought performance (Anderegg et al., 2018) but are time-consuming to measure and therefore infeasible for predicting or modeling drought responses in highly diverse forests (e.q., in the tropics). More easily measurable leaf hydraulic traits with direct linkage to plant hydraulic function can explain greater 97 variation in plant distribution and function (Medeiros et al., 2019). These include leaf area shrinkage upon desiccation (PLA_{dry}) (Scoffoni et al., 2014) and the leaf water potential at turgor loss point (π_{tlp}), i.e., the qq water potential at which leaf wilting occurs (Bartlett et al., 2016). The abilities of both PLA_{dry} and π_{tlp} to 100 explain tree performance under drought remains untested.

Here, we examine how tree height, microenvironment characteristics, and species' traits collectively shape 102 drought responses. We test a series of hypotheses and associated specific predictions (Table 1) based on the 103 combination of tree-ring records from three droughts (1966, 1977, 1999), species functional and hydraulic 104 trait measurements, and census data from a large forest dynamics plot in Virginia, USA. First, we focus on the role of tree size and its interaction with microenvironment. We examine the contemporary relationship 106 between tree height and microenvironment, including growing season meteorological conditions and crown 107 exposure. We then test whether, consistent with most forests globally, larger-diameter, taller trees tend to have lower drought resistance (Rt) in this forest, which is in a region (eastern North America) represented by 109 only two studies in the global review of Bennett et al. (2015). We also test for an influence of potential access 110 to available soil water, which should be greater for larger trees in dry but not in perpetually wet microsites. 111 Finally, we focus on the role of species' functional and hydraulic traits, testing the hypothesis that species' 112 traits—particularly leaf hydraulic traits—predict Rt. We test predictions that drought resistance is higher in 113 ring-porous than semi-ring and diffuse-porous species, that it is correlated with wood density-either postively 114 (Greenwood et al., 2017) or negatively (Hoffmann et al., 2011)—and positively correlated with LMA, and 115 that hydraulic leaf traits including PLA_{dry} and π_{tlp} are better predictors. 116

Materials and Methods

118 Study site and microclimate

Research was conducted at the 25.6-ha ForestGEO (Forest Global Earth Observatory) study plot at the Smithsonian Conservation Biology Institute (SCBI) in Virginia, USA (38°53'36.6"N, 78°08'43.4"W; Fig. S1)
(Bourg et al., 2013; Anderson-Teixeira et al., 2015a). SCBI is located in the central Appalachian Mountains near the northern boundary of Shenandoah National Park. Elevations range from 273 to 338 m above sea

level with a topographic relief of 65m (Bourg et al., 2013). Climate is humid temperate, with mean annual 123 temperature of 12.7°C and precipitation of 1005 mm during our study period (1960-2009; source: CRU TS v.4.01; Harris et al. (2014)). Dominant tree taxa within this secondary forest include Liriodendron tulipifera, 125 oaks (Quercus spp.), and hickories (Carya spp.; Table 2). 126 To characterize how environmental conditions vary with height, data were obtained from the NEON tower 127 located <1km from the study area via the neonUtilities package (?). We used wind speed, relative humidity, and air temperature data, all measured over a vertical profile spanning heights from 7.2 m to above the top of 129 the tree canopy (31.0 or 51.8m, depending on censor), for the years 2016-2018 (NEON, 2018). After filtering 130 for missing and outlier values, we determined the daily minima and maxima, which we then aggregated at the monthly scale. (work on organization of this.) 132 Identifying drought years 133 We identified the three largest droughts within the time period 1950-2009, defining drought (Slette et al., 134 2019) as events with anomalously dry peak growing season climatic conditions. Specifically, we used the 135 metric of Palmer Drought Severity Index (PDSI) during May-August (MJJA; Table S3), which were 136 identified by Helcoski et al. (2019) as the months of the current year to which annual tree growth was most 137 sensitive at this site. PDSI divisional data for Northern Virginia were obtained from NOAA (https://www7.ncdc.noaa.gov/CDO/CDODivisionalSelect.jsp) in December 2017. Based on this, we 139 identified three drought years - 1966, 1977, and 1999 (Figs. 1, S2, Table S3). 140 The droughts differed in intensity and antecedent moisture conditions (Fig. S2, Table S3). The 1966 drought 141 was preceded by two years of moderate drought during the growing season and severe to extreme drought starting the previous fall and in August reached the lowest growing season PDSI (-4.82) of the three 143 droughts. The 1977 drought was the least intense throughout the growing season, and it was preceded by 2.5 144 years of near-normal conditions, making it the mildest of the three droughts. The 1999 drought was preceded by wetter than average conditions until the previous June, but reached the lowest PDSI during May-July 146 (-4.53).147

148 Data collection and preparation

microenvironment characteristics, and species traits (Table 3). The SCBI ForestGEO plot was censused in 150 2008, 2013, and 2018 following standard ForestGEO protocols, whereby all free-standing woody stems > 1cm 151 diameter at breast height (DBH) were mapped, tagged, measured at DBH, and identified to species (Condit, 152 1998). From this census data, we used measurements of DBH from 2008 to calculate historical DBH and data 153 for all stems > 10cm to analyze functional trait composition relative to tree height (all analyses described 154 below). Census data are available through the ForestGEO data portal (www.forestgeo.si.edu). 155 We analyzed tree-ring data (xylem growth increment) from 571 trees representing the twelve species with the greatest contributions to woody aboveground net primary productivity $(ANPP_{stem})$, which together 157 comprised 97% of study plot ANPP_{stem} between 2008 and 2013 (Helcoski et al., 2019) (Table 2; Fig. S1). 158 Cores (one per tree) were collected within the ForestGEO plot at breast height (1.3m) in 2010-2011 or 159 2016-2017. In 2010-2011, cores were collected from randomly selected live trees of each species that had at 160 least 30 individuals ≥ 10 cm DBH (Bourg et al., 2013). In 2016-2017, cores were collected from all trees 161 found dead during annual mortality censuses (Gonzalez-Akre et al., 2016). We note that drought was

Within or just outside the ForestGEO plot, we collected data on a suite of variables including tree size,

probably not a cause of mortality for these trees, as monthly May-Aug PDSI did not drop below -1.75 in these years or the three years prior (2013-2017), and that trees cored dead displayed similar climate sensitivity to trees cored live (Helcoski et al., 2019). Cores were sanded, measured, and crossdated using 165 standard procedures, as detailed in (Helcoski et al., 2019). The resulting chronologies (Fig. 1a) were 166 published in Zenodo (DOI: 10.5281/zenodo.2649302) in association with Helcoski et al. (2019).

For each cored tree, we combined tree-ring records and allometric equations of bark thickness to reconstruct DBH for the years 1950-2009. Prior DBH was estimated using the following equation:

$$DBH_Y = DBH_{2008} - 2 * \left[\sum_{year=Y}^{2008} (r_{ring,Y}) - r_{bark,Y} + r_{bark,2008} \right]$$

Here, Y denotes the year of interest, r_{ring} denotes ring width derived from cores, and r_{bark} denotes bark thickness. Bark thickness was estimated from species-specific allometries based on the bark thickness data 171 from the site (Anderson-Teixeira et al., 2015b). Specifically, we used linear regression on log-transformed data to relate r_{bark} to diameter inside bark from 2008 data (Table S1), which were then used to determine 173 r_{bark} in the DBH reconstruction. 174 Tree heights (H) were measured by several researchers for a variety of purposes between 2012 to 2019 (n=1,518 trees). Measurement methods included direct measurements using a collapsible measurement rod on small trees (NEON, 2018) or a tape measure on recently fallen trees (this study); geometric calculations 177 using clinometer and tape measure (Stovall et al., 2018a) or digital rangefinders (Anderson-Teixeira et al., 178 2015b; NEON, 2018); and ground-based LiDAR (Stovall et al., 2018b). Rangefinders used either the tangent 179 method (Impulse 200LR, TruPulse 360R) or the sine method (Nikon ForestryPro) for calculating heights.

Both methods are associated with some error (Larjavaara and Muller-Landau, 2013), but in this instance 181 there was no clear advantage of one or the other. Measurements from the National Ecological Observatory Network (NEON) were collected near the ForestGEO plot following standard NEON protocol, whereby 183

vegetation of short stature was measured with a collapsible measurement rod, and taller trees with a

rangefinder (NEON, 2018). Species-specific height allometries were developed (Table S2) using logarithmic 185

regression $(ln[H] \sim ln[DBH])$. For species with insufficient height data to create reliable species-specific 186

allometries (n=2, JUNI and FRAM), heights were calculated from an equation developed by combining the height measurements across all species. We then used these allometries to estimate H for each drought year,

Y, based on reconstructed DBH_Y . 189

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Crown position—a categorical variable classifying trees based on exposure to sunlight—was recorded for all 190 cored trees that remained standing during the growing season of 2018 following the protocol of Jennings et al. (1999). Trees were classified as follows: dominant trees were defined as those with crowns above the general 192 level of the canopy, codominant trees as those with crowns within the the canopy; intermediate trees as those 193 with crowns below the canopy level, but illuminated from above; and suppressed as those below the canopy and receiving minimal direct illumination from above. 195

Topographic wetness index (TWI) was calculated using the dynatopmodel package in R (Fig. S1) (?). 196 Originally developed by Beven and Kirkby (1979), TWI was part of a hydrological run-off model and has 197 since been used for a number of purposes in hydrology and ecology (Sørensen et al., 2006). TWI calculation depends on an input of a digital elevation model (DEM; ~3.7 m resolution from the elevatr package (?)), and

from this yields a quantitative assessment defined by how "wet" an area is, based on areas where run-off is more likely. From our observations in the plot, TWI performed better at categorizing wet areas than the Euclidean distance from the stream. 202 Hydraulic traits were collected in August 2018 (Tables 2-3; Fig. S4). We sampled small sun-exposed 203 branches up to eight meters above ground from three individuals of each species in and around the 204 ForestGEO plot. Sampled branches were re-cut under water at least two nodes above the original cut and re-hydrated overnight in covered buckets under opaque plastic bags before measurements were taken. 206 Rehydrated leaves taken towards the apical end of the branch (n=3 per individual: small, medium, and 207 large) were scanned, weighed, dried at 60° C for ≥ 48 hours, and then re-scanned and weighed. Leaf area was calculated from scanned images using the LeafArea R package (Katabuchi, 2019). LMA was calculated 209 as the ratio of leaf dry mass to fresh area. PLA_{dry} was calculated as the percent loss of area between fresh 210 and dry leaves. Wood density was calculated for ~1cm diameter stem samples (bark and pith removed) as the ratio of dry weight to fresh volume, which was estimated using Archimedes' displacement. We used the 212 rapid determination method of Bartlett et al. (2012) to estimate osmotic potential at turgor loss point (π_{tlp}) . 213 Briefly, two 4 mm diameter leaf discs were cut from each leaf, tightly wrapped in foil, submerged in liquid 214 nitrogen, perforated 10-15 times with a dissection needle, and then measured using a vapour pressure 215 osmometer (VAPRO 5520, Wescor, Logan, UT, USA). Osmotic potential (π_{osm}) given by the osmometer was 216 used to estimate (π_{tlp}) using the equation $\pi_{tlp} = 0.832\pi_{osm}^{-0.631}$ (Bartlett et al., 2012). 217 Statistical Analysis 218 For each drought year, we calculated drought resistance (Rt) as the ratio of basal area increment (BAI)219 during drought to the mean BAI over the five years preceding the drought (Lloret et al., 2011). Thus, Rt220 values <1 and >1 indicate growth reductions and increases, respectively. Because the Rt metric could be 221 biased by directional pre-drought growth trends, we also tried an intervention time series analysis (ARIMA, 222 (?)) that predicted mean drought-year growth based on trends over the past ten years and used this value in 223 place of the five-year mean in calculations of resistance (Rt_{ARIMA} = observed BAI/ predicted BAI). The 224 two metrics were strongly correlated (Fig. S5). Because Rt tended to produce more reasonable estimates 225 than Rt_{ARIMA} when there was a large difference between these metrics, we selected Rt as our focal metric, 226 presenting parallel results for Rt_{ARIMA} in the Supplementary Info. We focus exclusively on drought 227 resistance (Rt or Rt_{ARIMA}), and not on the resilience metrics described in Lloret et al. (2011), because (1) 228 we would expect resilience to be controlled by a different set of mechanisms, and (2) the findings of DeSoto et 229 al. (2020) suggest that Rt is a more important drought response metric for angiosperms. Analyses focused on testing the predictions presented in Table 1, with Rt (or Rt_{ARIMA}) as the response 231 variable. Models were run for all drought years combined and for each drought year individually. The general 232 statistical model for hypothesis testing was a mixed effects model, implemented in the lme4 package in R 233 $[\mathbf{REF}]$, with Rt (or Rt_{ARIMA}) as the response variable, tree nested within species as a random effect, and 234 independent variables as specified below. In the multi-year model only, we also included a fixed effect of 235 drought year. (IAN, CONFIRM THIS)) We used AICc to assess model selection, and 236 conditional/marginal R-squared to assess model fit as implemented in the AICcmodavg package in R (?). AICc refers to a corrected version of AICc, and is best suited for small data sizes (see Brewer et al., 2016). 238

To avoid over-fitting models with five species traits (Table 3) across only 12 species, we did not include all

traits as fixed effects in a single GLMM, but rather conducted individual tests of each species trait to

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determine the relative importance and appropriateness for inclusion in the main model. These tests followed the model structure specified above, including ln[H], TWI, and crown position in the null model. Trait variables were considered appropriate for inclusion in the main model if they had a consistent direction of response across all droughts and if their addition to this null model lacking the trait improved fit (at $\Delta AICc$ ≥ 1.0) in at least one drought year (Table S4). We note that the $\Delta AICc \geq 1.0$ criterion is not a test of significance, but of whether the variable has enough influence to be considered as a *candidate* variable in full models.

We then determined the top full models for predicting Rt (or Rt_{ARIMA}). To do so, we compared models with all possible combinations of candidate variables, including ln[H]*TWI and species traits as specified above. We identified the full set of models within $\Delta AICc=2$ of the best model (that with lowest AICc). When a variable appeared in all of these models and the sign of the coefficient was consistent across models, we viewed this as support for the acceptance/rejection of the associated prediction (Table 1). If the variable appeared in some but not all of these models, and its sign we was consistent across models, we considered this partial support/rejection. In presentation of the results below, we note instances where the Rt_{ARIMA} model disagreed with the Rt model, but otherwise do not discuss the Rt_{ARIMA} model.

All analysis beyond basic data collection was performed using R version 3.5.3 (R Core Team, 2020). Other R-packages aside from those already listed were very helpful in conducting analyses. These are listed in the Supplementary Information. All data, code, and results are available through the SCBI-ForestGEO organization on GitHub (https://github.com/SCBI-ForestGEO: SCBI-ForestGEO-Data and McGregor_climate-sensitivity-variation repositories), with static versions corresponding to data and analyses presented here archived in Zenodo (DOIs: 10.5281/zenodo.3604993 and [TBD], respectively.

262 Results

263 Tree height and microenvironment

In the years for which we have vertical profiles in climate data (2016-2018), taller trees—or those in dominant crown positions— were generally exposed to higher evaporative demand during the peak growing season months (May-August; Fig. 3). Specifically, maximum daily wind speeds were significantly higher above the top of the canopy (40-50m) than within and below (10-30m) (Fig. 3a). Relative humidity was also somewhat lower during June-August, ranging from ~50-80% above the canopy and ~60-90% in the understory (Fig. 3b). Air temperature did not vary across the vertical profile (Fig. 3c).

270 Crown position varied as expected with height (dominant > co-dominant > intermediate > suppressed), but 271 with substantial variation (Fig. 3d). Correlation between height and canopy position was 0.73.

272 Community-level and species' drought responses

At the community level, cored trees showed substantial growth reductions in all three droughts, with a mean Rt of 0.86 in 1966 and 1999, and 0.84 in 1977 (Fig. 1b). Across the entire study period (1950-2009), the focal drought years were the three years with the largest fraction of trees exhibiting $Rt \leq 0.7$. Specifically, in each drought, roughly 30% of the cored trees had growth reductions of $\geq 30\%$ ($Rt \leq 0.7$): 29% in 1966, 32% in 1977, and 27% in 1999. However, some individuals exhibited increased growth, *i.e.*, Rt > 1.0: 26% of trees in 1966, 22% in 1977, and 26% in 1999.

Responses varied across species and by drought (Fig. 2). Averaged across all droughts, Rt was lowest in

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Tree height, microenvironment, and drought resistance
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    Taller trees showed stronger growth reductions during drought when evaluating the three drought years
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    together and for 1966 individually (Table 1; Fig. 4). Specifically, ln[H] appeared, with negative coefficient,
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    in the best models and all top models (Tables S6-S7). For the 1977 drought, ln[H] did not appear in the best
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    model, but was included, with negative coefficient, among the top models-i.e., models that were statistically
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    indistinguishable (\triangle AICc < 2) from the best model (Tables 1, S6-S7). For the 1999 drought, ln[H] appeared
    in no top models for Rt and some top models, with positive coefficient, for Rt_{ARIMA}
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    Rt had a significantly negative response to ln[TWI] in all drought years combined and in 1977 and 1999
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    individually (Fig. 4, Table S6). When Rt_{ARIMA} was used as the response variable, the effect was significant
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    in all drought years combined and in 1977, and a negative effect of ln[TWI] was included in some of the
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    models in 1966 and 1999 (Table S7). This negates the idea that trees in moist microsites would be less
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    affected by drought. Nevertheless, we tested for a ln[H] * ln[TWI] interaction, a negative sign of which could
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    indicate that smaller trees (with smaller rooting volume) are more susceptible to drought in drier
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    microenvironments with a deeper water table. This hypothesis was rejected, as the ln[H] * ln[TWI]
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    interaction was never significant, and had a positive sign in any top models in which it appeared (Tables 1.
    S6-S7). This term did appear, with positive coefficient, in the best Rt_{ARIMA} model for all years combined
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    Species' traits and drought resistance
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    Species traits... (Table 2, Fig. S4)
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    Wood density, LMA, and xylem porosity were all poor predictors of Rt (Tables 1,S4-S5). Wood density and
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    LMA were never significantly associated with Rt in the single-variable tests and were therefore excluded
    from the full models. Xylem porosity was also excluded from the full models, as it had no significant
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    influence for all droughts combined and had contrasting effects in the individual droughts: whereas
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    ring-porous species had higher Rt than diffuse- and semi-ring- porous species in the 1966 and 1999 droughts,
    they had lower Rt in 1977 (Table S4). It is noteworthy that the two diffuse-porous species in our study,
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    Liriodendron tulipifera and Faqus grandifolia, were at opposite ends of the Rt spectrum (Fig. 2), further
    refuting the idea that xylem porosity is a useful predictor of Rt in the context of this study.
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    In contrast, PLA_{dry}, and \pi_{tlp} were linked to drought responses (Fig. 4; Tables 1,S4-S7). Both had consistent
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    signs across all droughts and explained modest amounts of variation (\Delta AICc > 1.0) during at least one of
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    the three droughts (Table S4), qualifying them as candidate variables for the full model. PLA_{dry} had a
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    significant influence, with negative coefficient, in full models for the three droughts combined and for the
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    1966 drought individually (Fig. 4; Tables S6-S7). For 1977 and 1999, it was included, with negative
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    coefficient, in some of the top models (Tables S6-S7). \pi_{tlp} was included, with negative coefficient, in the best
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    model for all droughts combined and for the 1977 drought individually (Fig. 4: Table S6). It was included in
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    some of the top models for 1999 (Tables S6-S7).
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Liriodendron tulipifera (mean Rt = 0.66) and highest in Fagus grandifolia (mean Rt = 0.99).

316 Discussion

Tree height, microenvironment, and hydraulic traits shaped tree growth responses across three droughts at our study site (Table 1, Fig. 4). The greater susceptibility of larger trees to drought, similar to forests

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worldwide (Bennett et al., 2015), was driven primarily by their height (Stovall et al., 2019). Taller height was
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    likely a liability in itself, and was also associated with greater exposure to conditions that would promote
    water loss and heat damage during drought (Fig. 3). There was no evidence that soil water availability
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    increased drought resistance; in contrast, trees in wetter topographic positions had lower Rt (Zuleta et al.,
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    2017; Stovall et al., 2019), and the larger potential rooting volume of large trees provided no advantage in the
    drier microenvironments. The negative effect of height on Rt held after accounting for species' traits.
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    Drought resistance was not consistently linked to species' LMA, wood density, or xylem type (ring-
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    vs. diffuse porous), but was negatively correlated with leaf hydraulic traits (PLA_{dry}, \pi_{tlp}). This is the first
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    report to our knowledge linking PLA_{dry} and \pi_{tlp} to growth reduction during drought. The directions of
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    these responses were consistent across droughts (Table S6), supporting the premise that they were driven by
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    fundamental physiological mechanisms. However, the strengths of each predictor varied across droughts (Fig.
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    4; Tables S6-S7), indicating that drought characteristics interact with tree size, microenvironment, and traits
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    to shape which individuals are most affected. These findings advance our knowledge of the factors that make
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    trees vulnerable to growth declines during drought-and, by extension, likely make them more vulnerable to
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    mortality (Sapes et al., 2019).
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    The droughts considered here were of a magnitude that has occurred with an average frequency of
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    approximately once every 10-15 years (Fig. 1a, Helcoski et al. (2019)) and had substantial but not
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    devastating impacts on tree growth (Figs. 1b, 2). These droughts were classified as severe (1977) or extreme
    (1966, 1999) according to the PDSI metric and have been linked to tree mortality in the eastern United
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    States (Druckenbrod et al., 2019); however, extreme, multiannual droughts or so-called "megadroughts" of
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    the type that have triggered massive tree die-off in other regions (e.g., Allen et al. (2010); Stovall et al.
339
    (2019)) have not occurred in the Eastern United States within the past several decades (Clark et al., 2016).
340
    Of the droughts considered here, the 1966 drought, which was preceded by two years of dry conditions (Fig.
341
    S2), severely stressed a larger portion of trees (Fig. 1b). The tendency for large trees to have lowest
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    resistance was most pronounced in this drought, consistent with other findings that this physiological
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    response increases with drought severity (Bennett et al., 2015; Stovall et al., 2019). Across all three droughts,
344
    the majority of trees experienced reduced growth, but a substantial portion had increased growth (Fig. 1b),
    potentially due to decreased leaf area of competitors during the drought (REF-if we can find one), and
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    consistent with prior observations that smaller trees can exhibit increased growth rates during drought
347
    (Bennett et al., 2015). It is likely because of the moderate impact of these droughts, along with other factors
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    influencing tree growth (e.g., stand dynamics), that our best models characterize only a modest amount of
349
    variation in Rt: 11-12\% for all droughts combined, and 18-25\% for each individual drought (Table S6).
350
    Consistent with studies in other forests worldwide (REFS), taller trees in this forest exhibited lower drought
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    resistance. Mechanistically, this is consistent with, and reinforces, previous findings that biophysical
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    constraints make it impossible for trees to efficiently transport water to great heights and simultaneously
353
    maintain strong resistance and resilience to drought-induced embolism (Olson et al., 2018; Couvreur et al.,
354
    2018; Roskilly et al., 2019). Taller trees also face dramatically different microenvironments (Fig. 3). They
355
    are exposed to higher wind speeds and lower humidity (Fig. 3a-b), resulting in higher evaporative demand
356
    (REFS). Unlike other temperate forests where modestly cooler understory conditions have been documented
357
    (REFS), particularly under drier conditions (REFS), we observed no significant variation in air
358
    temperatures across the vertical profile (Fig. 3c). More critically for tree physiology, leaf temperatures can
359
    become significantly elevated over air temperature under conditions of high solar radiation and low stomatal
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conductance (Campbell & Norman REF). Under drought, when air temperatures tend to be warmer, direct 361 solar radiation tends to be higher (because of less cloud cover), and less water is available for evaporative cooling of the leaves, trees with sun-exposed crowns may not be able to simultaneously maintain leaf 363 temperatures below damaging extremes and avoid drought-induced embolism. Indeed previous studies have 364 shown lower drought resistance in more exposed trees (Liu and Muller, 1993; Suarez et al., 2004; Scharnweber et al., 2019). Unfortunately, collinearity between height and crown exposure in this study (Fig. 366 3d) makes it impossible to confidently partition causality. Additional research comparing drought responses 367 of early successional and mature forest stands, along with short and tall isolated trees, would be valuable for more clearly disentangling the roles of tree height and crown exposure. 369 Belowground, taller trees would tend to have larger root systems, but the potentially greater access to water 370 did not override the disadvantage conferred by height-and, in fact, greater moisture access in non-drought years (here, higher TWI) appears to make trees more sensitive to drought (Zuleta et al., 2017; Stovall et al., 372 2019). The observed height-sensitivity of Rt, together with the lack of advantage to large stature in drier 373 topographic positions, agrees with the concept that physiological limitations to transpiration under drought 374 shift from soil water availability to the plant-atmosphere interface as forests age (Bretfeld et al., 2018), such 375 that tall, dominant trees are the most sensitive in mature forests. Again, additional research comparing 376 drought responses across forests with different tree heights and water availability would be valuable for 377 disentangling the relative importance of above- and belowground mechanisms. 378 Even under non-drought conditions, evaporative demand and maximum leaf temperatures increase with tree 379 height (Smith and Nobel, 1977; Bretfeld et al., 2018; Kunert et al., 2017), and such conditions would incur 380 additional stress during drought, when solar radiation tends to be higher and less water is available for 381 evaporative cooling of the leaves. 382 The development of tree-ring chronologies for the twelve most dominant tree species at our site (Helcoski 383 et al., 2019; Bourg et al., 2013) gave us the sample size to compare historical drought responses across 384 species (Fig. 2) and associated traits at a single site (see also Elliott et al., 2015). Our study reinforced the 385 findings of previous studies (see Introduction) that wood density and LMA are not reliably linked to drought 386 resistance (Table 1). Contrary to previous studies in temperate deciduous forests, we did not find an 387 association between xylem porosity and drought tolerance, as the two diffuse-porous species, Liriodendron 388 tulipifera and Fagus grandifolia, were at opposite ends of the Rt spectrum (Fig. 2). While the low Rt of L. 389 tulipifera is consistent with other studies (Elliott et al., 2015), the high Rt of F. grandifolia contrasts with studies identifying diffuse porous species in general (Elliott et al., 2015; Kannenberg et al., 2019), and the 391 genus Faqus in particular (Friedrichs et al., 2009), as drought sensitive. 392 There are two potential explanations for this discrepancy. First, other traits can and do override the influence of xylem porosity on drought resistance. Ring porous species are restricted mainly to temperate deciduous 394 forests (Wheeler et al. 2007), while highly drought-tolerant diffuse-porous species exist in other biomes 395 (REFS). Fagus grandifolia had intermediate π_{tlp} and low PLA_{dry} (Fig. S4), which would have contributed to it's drought resistance (Fig. 4; see discussion below). A second explanation of why F. grandifolia trees at 397 this particular site had higher Rt is that the sampled individuals, reflective of the population within the plot, 398 are generally shorter and in less dominant canopy positions compared to most other species (Fig. S4). The 399 species, which is highly shade-tolerant, also has deep crowns (Anderson-Teixeira et al., 2015b), implying that 400 a lower proportion of leaves would be affected by harsher microclimatic conditions at the top of the canopy 401 under drought (Fig. 3). Thus, the high Rt of the sampled F. grandifolia population can be explained by a

combination of fairly drought-resistant leaf traits, shorter stature, and a buffered microenvironment. 403 Concerted measurement of tree-rings and leaf hydraulic traits of emerging importance (Scoffoni et al., 2014; 404 Bartlett et al., 2016; Medeiros et al., 2019) allowed novel insights into the role of hydraulic traits in shaping drought response. The finding that PLA_{dry} and π_{tlp} can be useful for predicting drought responses of tree 406 growth (Fig. 4; Table 1) is both novel and consistent with previous studies linking these traits to habitat and 407 drought tolerance. Previous studies have demonstrated that π_{tlp} and PLA_{dry} are physiologically meaningful traits linked to species distribution along moisture gradients (Maréchaux et al., 2015; Fletcher et al., 2018; 409 Medeiros et al., 2019; Simeone et al., 2019; Rosas et al., 2019), and our findings indicate that these traits also 410 influence drought responses. Furthermore, the observed linkage of π_{tlp} to Rt in this forest aligns with observations in the Amazon that π_{tlp} is higher in drought-intolerant than drought-tolerant plant functional 412 types and adds support to the idea that this trait is useful for categorizing and representing species' drought 413 responses in models (Powell et al., 2017). Because both PLA_{dry} and π_{tlp} can be measured relatively easily (Bartlett et al., 2012; Scoffoni et al., 2014), they hold promise for predicting drought growth responses across 415 diverse forests. The importance of predicting drought responses from species traits increases with tree species 416 diversity; whereas it is feasible to study drought responses for all dominant species in most boreal and 417 temperate forests (e.g., this study), this becomes difficult to impossible for species that do not form annual 418 rings, and for diverse tropical forests. Although progress is being made for the tropics (Schöngart et al., 419 2017), a full linkage of hydraulic traits to drought responses would be invaluable for forecasting how little-known species and whole forests will respond to future droughts (Powell et al., 2017). 421 As climate change drives increasing drought in many of the world's forests (Trenberth et al., 2014; 422 Intergovernmental Panel on Climate Change, 2015), the fate of forests and their climate feedbacks will be 423 shaped by the biophysical and physiological drivers observed here. Large trees have been disproportionately 424 impacted by strong drought in forests around the world (Bennett et al., 2015; Stovall et al., 2019), and we 425 show, at least at this site, that this is primarily driven by their height, potentially with some contributions 426 from crown position. The distinction is important because it suggests that height per se makes trees 427 vulnerable, even if their crowns are somewhat protected by neighbors, whereas shorter solitary trees or the 428 dominant trees in young forests that recently established after logging or natural disturbances should be less 429 vulnerable. This would suggest that, all else being equal, mature forests would be more vulnerable to 430 drought than young forests with short trees; however, root water access may limit the young forests (Bretfeld 431 et al., 2018), and species traits often shift as forests age. Early- to mid- successional species at our site (Liriodendron tulipifera, Quercus spp., Fraxinus americana) display a mix of traits conferring drought 433 tolerance and resistance (Table 2), and further research on how hydraulic traits and drought vulnerability 434 change over the course of succession would be valuable for addressing how drought tolerance changes as 435 forests age (e.g. Rodríguez-Catón et al., 2015). In the meantime, the results of this study advance our knowledge of the factors conferring drought resistance in a mature forest, opening the door for more accurate 437

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forecasting of forest responses to future drought.

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450 Author Contribution

- 451 KAT, IM, and AJT designed the research. Tree-ring chronologies were developed by RH under guidance of
- 452 AJT and NP. Trait data was collected by IM, JZ under guidance of NK and LS. Other plot data were
- collected by IM, AS, EGA, and NB under guidance of EGA and WM. Data analyses were performed by IM
- under guidance of KAT and VH. KAT and IM interpreted the results. IM and KAT wrote the first draft of
- manuscript, and all authors contributed to revisions.

Supplementary Information

- Table S1: Species-specific bark thickness regression equations
- Table S2: Species-specific height regression equations
- Table S3: Palmer drought severity index (PDSI) by month for focal droughts
- Figure S1: Map of ForestGEO plot showing TWI and location of cored trees
- Figure S2: Time series of Palmer Drought Severity Index (PDSI) for the 2.5 years prior to each focal drought
- 462 Figure S3: Height (from reconstructed DBH) by crown position across the three focal droughts and in the
- year of measurement (2018)

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